

Kolmogorov numberings and minimal identification

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Abstract

Identification of programs for computable functions from their graphs by algorithmic devices is a well studied problem in learning theory. Freivalds and Chen consider identification of ‘minimal’ and ‘nearly minimal’ programs for functions from their graphs. To address certain problems in minimal identification for Gödel numberings, Freivalds later considered minimal identification in Kolmogorov numberings. Kolmogorov numberings are in some sense optimal numberings and have some nice properties. We prove certain separation results for minimal identification in every Kolmogorov numbering. In addition we also compare minimal identification in Gödel numberings versus minimal identification in Kolmogorov numberings.

1. Introduction

Suppose f is a total recursive function. For any natural number n , we let $f[n]$ denote $\{(x, f(x)) \mid x < n\}$, the finite initial segment of f consisting of the first n data points in the graph of f . Criteria of inference informally described below are formally defined in Section 3. In this paper we are only concerned about learnability of total recursive functions.

A *function learning machine* \mathbf{M} is an algorithmic device which, on any input segment $f[n]$, returns either ? or a program. The output of \mathbf{M} on $f[n]$ is denoted by $\mathbf{M}(f[n])$. If $\mathbf{M}(f[n])$ is a program, we think of that program as \mathbf{M} ’s conjecture, based on the data $f[n]$, about how to compute all of f ; $\mathbf{M}(f[n]) = ?$ then represents the situation where \mathbf{M} does not conjecture a program based on the data $f[n]$.

As is by now well known, there are various senses in which \mathbf{M} can be thought of as *successfully* learning or inferring a program for f . Let $p_n = \mathbf{M}(f[n])$. The criterion of success known as *Ex-identification* [2, 4, 11] requires that the sequence p_0, p_1, p_2, \dots

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contains a program p , which computes f , such that, for all but finitely many i , $p_i = p$. In this case one speaks of p as being the *final* program output by \mathbf{M} on f .

Based on the principle of Occam's Razor, Freivalds [7] and Chen [5, 6] studied the effect of requiring that the final hypothesis held by the learner in the above model be of (nearly) minimal size. Suppose ψ is a numbering (programming system). In $\mathbf{Min}_\psi\mathbf{Ex}$ -identification criterion one requires, in addition to \mathbf{Ex} -identification of function (in the programming system ψ), that the final program be of minimal size (see formal definitions in Section 3). Minimal identification of classes of functions depends on the acceptable programming system (acceptable numbering) chosen to interpret programs output by learning machines. We direct the reader to [5, 6, 7] for results dealing with minimal identification and its relationship with \mathbf{Ex} -identification.

Freivalds [7] showed that there are Gödel numberings in which no function class of infinite cardinality can be learned by minimal programs. This led Freivalds to consider minimal identification in Kolmogorov numberings. Kolmogorov numberings are numberings to which every other numbering is reducible by a linearly bounded function. Freivalds showed that for every Kolmogorov numbering ψ , there exists an infinite class of functions in $\mathbf{Min}_\psi\mathbf{Ex}$. In this paper we consider the relationship between identification criteria such as $\mathbf{Min}_\psi\mathbf{Ex}$, $\mathbf{Min}_\psi\mathbf{FIN}$ and $\mathbf{Min}_\psi\mathbf{CoLearn}$ in Kolmogorov numberings ψ (see formal definitions in Section 3). We show that these criteria are separated in every Kolmogorov numbering. We also show the existence of classes of functions which can be minimally identified in some Gödel numbering but cannot be minimally identified in any Kolmogorov numbering.

We now proceed formally.

2. Notation

Recursion-theoretic concepts not explained below are treated in [14]. N denotes the set of natural numbers, $\{0, 1, 2, \dots\}$. The symbols $c, e, i, j, k, l, m, n, p, r, s, t, u, v, x, y$ and z , with or without decorations (decorations are subscripts, superscripts and the like), range over natural numbers unless otherwise specified. $\subseteq, \subset, \supseteq, \supset, \in$, denote subset, proper subset, superset, proper superset and membership relationship respectively. \emptyset denotes the empty set. A, C, L, S , with or without decorations, range over subsets of N . \bar{L} denotes the complement of L , i.e. $\bar{L} = N - L$. We denote the cardinality of a set S by $\text{card}(S)$. $\max(\cdot), \min(\cdot)$ denote the maximum and minimum of a set, respectively. By convention $\max(\emptyset) = 0$ and $\min(\emptyset) = \infty$. The quantifiers \forall^∞ and \exists^∞ mean 'for all but finitely many' and 'there exist infinitely many', respectively.

\mathcal{R} denotes the set of all total recursive functions. h, f, g, q , with or without decorations, range over \mathcal{R} . \mathcal{C} and \mathcal{S} , with or without decorations, range over subsets of \mathcal{R} . \downarrow denotes defined. \uparrow denotes undefined. ξ , with or without decorations, ranges over partial recursive functions.

A programming system (or computable numbering) is a (partial) computable function of two variables. We let ψ, β, η range over computable numberings (programming

systems). Suppose $\psi(\cdot, \cdot)$ is a computable numbering. We often refer to the (partial) function $\lambda x. \psi(i, x)$ as ψ_i . ψ_i thus denotes the (partial) function computed by the i th program in the numbering ψ . Note that, in general, a computable numbering may not contain all the partial recursive functions. We often drop the word ‘computable’ from ‘computable numbering’ in this paper, since we will be dealing with computable numberings only.

An acceptable numbering is a computable numbering to which every other computable numbering is reducible via a recursive function. Thus if ψ is an acceptable numbering, then for all computable numberings η , there exists a recursive function h such that $(\forall i)[\eta_i = \psi_{h(i)}]$. Acceptable numberings are also referred to as Gödel numberings. Kolmogorov numbering is an acceptable numbering to which every other computable numbering can be reduced via a linearly bounded function. Thus if ψ is a Kolmogorov numbering, then for all computable numberings η , there exists a recursive function h and a constant c such that $(\forall i)[\eta_i = \psi_{h(i)} \wedge h(i) \leq \max(\{c * i, c\})]$.

For a function f , $\text{MinProg}_\psi(f)$ denotes the minimal program for f , if any, in the ψ programming system, i.e., $\text{MinProg}_\psi(f) = \min(\{i \mid \psi_i = f\})$. Let $\mathbf{ZEROSTAR} = \{f \mid (\forall^\infty x)[f(x) = 0]\}$.

We let φ denote an arbitrary fixed acceptable programming system. φ_i thus denotes the partial recursive function computed by the i th program in the acceptable programming system φ . We often refer to the i th program as program i . Φ denotes an arbitrary fixed Blum complexity measure [1, 12] for the φ -system.

$\lambda i, j. \langle i, j \rangle$ stands for an arbitrary computable one to one encoding of all pairs of natural numbers onto N [14]. We assume that the pairing function is such that $\langle i, j \rangle \geq \max(\{i, j\})$.

3. Learning paradigms

For any partial function ξ and any natural number n such that, for each $x < n$, $\xi(x) \downarrow$, we let $\xi[n]$ denote the finite initial segment $\{(x, \xi(x)) \mid x < n\}$. Let $\text{INIT} = \{f[n] \mid f \in \mathcal{R} \wedge n \in N\}$. We let σ, τ and γ , with or without decorations, range over INIT . We let Λ denote the empty sequence. $|\sigma|$ denotes the length of σ . Thus for example $|f[n]| = n$.

Suppose ξ is a partial function. Then $\text{zeroext}(\xi)$ denotes a function such that

$$(\text{zeroext}(\xi))(x) = \begin{cases} \xi(x) & \text{if } \xi(x) \downarrow, \\ 0 & \text{otherwise.} \end{cases}$$

Definition 1 (Gold [11]). A *learning machine* is an algorithmic device which computes a mapping from INIT into $N \cup \{?\}$ such that, if $\mathbf{M}(f[n]) \neq ?$, then $\mathbf{M}(f[n+1]) \neq ?$.

In Definition 1 above, ‘?’ denotes the situation when \mathbf{M} outputs ‘no conjecture’ on some $\sigma \in \text{INIT}$. The restriction that \mathbf{M} must continue to conjecture programs once it has done so is essentially without loss of generality since a machine which has not had

enough time to think of a new conjecture can be thought of re-outputting its previous conjecture.

We let \mathbf{M} , with or without superscripts, range over learning machines. (We reserve \mathbf{M} with subscripts for special type of enumeration of learning machines. See Proposition 1 below.)

Definition 2. Suppose \mathbf{M} is a learning machine and f is a computable function. $\mathbf{M}(f) \downarrow$ (read: $\mathbf{M}(f)$ converges) just in case $(\exists i)(\forall^\infty n) [\mathbf{M}(f[n]) = i]$. If $\mathbf{M}(f) \downarrow$, then $\mathbf{M}(f)$ is defined as the unique i such that $(\forall^\infty n) [\mathbf{M}(f[n]) = i]$, otherwise we say that $\mathbf{M}(f)$ diverges (written: $\mathbf{M}(f) \uparrow$).

We now formally define the criteria of inference considered in this paper.

3.1. Explanatory function identification

Definition 3 (Gold [11], Case and Smith [4]). (a) A learning machine \mathbf{M} is said to **Ex-identify** f (written: $f \in \mathbf{Ex}(\mathbf{M})$) just in case $(\exists i \mid \varphi_i = f) [\mathbf{M}(f) \downarrow \wedge \mathbf{M}(f) = i]$.

(b) $\mathbf{Ex} = \{\mathcal{C} \mid (\exists \mathbf{M}) [\mathcal{C} \subseteq \mathbf{Ex}(\mathbf{M})]\}$.

3.2. Finite function inference

Definition 4 (Gold [11]). (a) A learning machine \mathbf{M} is said to **FIN-identify** f (written: $f \in \mathbf{FIN}(\mathbf{M})$) just in case $(\exists n, p \mid \varphi_p = f) [(\forall m < n) [\mathbf{M}(f[m]) = ?] \wedge (\forall m \geq n) [\mathbf{M}(f[m]) = p]]$.

(b) $\mathbf{FIN} = \{\mathcal{C} \mid (\exists \mathbf{M}) [\mathcal{C} \subseteq \mathbf{FIN}(\mathbf{M})]\}$.

3.3. Colearnability

We say that $\mathbf{M}(f)$ co-converges to p , iff $[N - \{\mathbf{M}(f[m]) \mid m \in N\} = \{p\}]$. If there exists a p such that $\mathbf{M}(f)$ co-converges to p , then we say that \mathbf{M} co-converges on f (to p). Otherwise we say that \mathbf{M} co-diverges on f .

Definition 5 (Freivalds et al. [10]). (a) A learning machine \mathbf{M} is said to **CoLearn** f (written: $f \in \mathbf{CoLearn}(\mathbf{M})$) just in case $(\exists p \mid \varphi_p = f) [\mathbf{M} \text{ co-converges on } f \text{ to } p]$.

(b) $\mathbf{CoLearn} = \{\mathcal{C} \mid (\exists \mathbf{M}) [\mathcal{C} \subseteq \mathbf{CoLearn}(\mathbf{M})]\}$.

For the study of colearnability it is useful to define the following notation. Suppose $\sigma, \tau \in \text{INIT}$. Then we define

$$P_{\mathbf{M}}(\sigma, \tau) = \{\mathbf{M}(\gamma) \mid \mathbf{M}(\gamma) \neq ? \wedge \sigma \subseteq \gamma \subseteq \tau\}.$$

Intuitively $P_{\mathbf{M}}(\sigma, \tau)$ denotes the set of programs output by \mathbf{M} on initial segments in $\{\gamma \mid \sigma \subseteq \gamma \subseteq \tau\}$.

Similarly, we define

$$P_{\mathbf{M}}(f[n], f) = \{\mathbf{M}(f[n']) \mid \mathbf{M}(f[n']) \neq ? \wedge n \leq n'\}.$$

3.4. Minimal identification

We next consider identification by minimal programs. Minimal identification usually depends on the numbering system chosen. Note that **Ex**, **CoLearn** and **FIN** are independent of acceptable programming system used. Thus, without loss of generality, we have used the φ acceptable programming system in the above definitions. However, minimal identification is acceptable programming system dependent, and thus we need to indicate the programming system in the definitions. For simplicity of presenting the proofs, we need to consider identification in non-acceptable programming systems also. Thus we consider a general definition of minimal identification, where the programming system used need not be acceptable.

Definition 6 (Freivalds [7]). Suppose ψ is a numbering.

- (a) \mathbf{M} is said to **Min $_{\psi}$ Ex-identify** f (written $f \in \mathbf{Min}_{\psi}\mathbf{Ex}(\mathbf{M})$) iff $\mathbf{M}(f) \downarrow \wedge \mathbf{M}(f) = \mathbf{MinProg}_{\psi}(f)$.
- (b) $\mathbf{Min}_{\psi}\mathbf{Ex} = \{\mathcal{C} \mid (\exists \mathbf{M})[\mathcal{C} \subseteq \mathbf{Min}_{\psi}\mathbf{Ex}(\mathbf{M})]\}$.
- (c) \mathbf{M} is said to **Min $_{\psi}$ FIN-identify** f (written $f \in \mathbf{Min}_{\psi}\mathbf{FIN}(\mathbf{M})$) iff $(\exists n)[(\forall m < n)[\mathbf{M}(f[n]) = ?] \wedge (\forall m \geq n)[\mathbf{M}(f[n]) = \mathbf{MinProg}_{\psi}(f)]]$.
- (d) $\mathbf{Min}_{\psi}\mathbf{FIN} = \{\mathcal{C} \mid (\exists \mathbf{M})[\mathcal{C} \subseteq \mathbf{Min}_{\psi}\mathbf{FIN}(\mathbf{M})]\}$.
- (e) \mathbf{M} is said to **Min $_{\psi}$ CoLearn** f (written $f \in \mathbf{Min}_{\psi}\mathbf{CoLearn}(\mathbf{M})$) iff $\mathbf{M}(f)$ co-converges to $\mathbf{MinProg}_{\psi}(f)$.
- (f) $\mathbf{Min}_{\psi}\mathbf{CoLearn} = \{\mathcal{C} \mid (\exists \mathbf{M})[\mathcal{C} \subseteq \mathbf{Min}_{\psi}\mathbf{CoLearn}(\mathbf{M})]\}$.

The following proposition facilitates the proof of some of our results.

Proposition 1. *There exists a recursively enumerable sequence $\mathbf{M}_0, \mathbf{M}_1, \dots$ of learning machines such that, for all machines \mathbf{M} and computable numberings ψ , there exists an i such that, for $\mathbf{I} \in \{\mathbf{Ex}, \mathbf{FIN}, \mathbf{CoLearn}, \mathbf{Min}_{\psi}\mathbf{Ex}, \mathbf{Min}_{\psi}\mathbf{FIN}, \mathbf{Min}_{\psi}\mathbf{CoLearn}\}$*

$$\mathbf{I}(\mathbf{M}) \subseteq \mathbf{I}(\mathbf{M}_i).$$

For a proof of the above proposition see for example [13]. We let $\mathbf{M}_0, \mathbf{M}_1, \dots$ be one such enumeration.

FINITE denotes the collection of all finite classes of total recursive functions, i.e. $\mathbf{FINITE} = \{\mathcal{C} \mid \mathcal{C} \subseteq \mathcal{R} \wedge \text{card}(\mathcal{C}) < \infty\}$.

4. Results

4.1. Relationship between minimal inference classes in different Kolmogorov numberings

It is easy to see that for all acceptable numberings ψ , $\mathbf{FINITE} \subseteq \mathbf{Min}_{\psi}\mathbf{FIN} \subseteq \mathbf{Min}_{\psi}\mathbf{CoLearn} \subseteq \mathbf{Min}_{\psi}\mathbf{Ex}$. In this section we show that **FINITE**, $\mathbf{Min}_{\psi}\mathbf{FIN}$, $\mathbf{Min}_{\psi}\mathbf{CoLearn}$, $\mathbf{Min}_{\psi}\mathbf{Ex}$ differ for each Kolmogorov numbering ψ . Note that this is

not the case for Gödel numberings since there exists a Gödel numbering η such that no infinite class of functions is in $\mathbf{Min}_\eta \mathbf{Ex}$.

Theorem 1 shows that $\mathbf{Min}_\psi \mathbf{FIN}$ and \mathbf{FINITE} differ for all Kolmogorov numberings; Theorem 2 shows that $\mathbf{Min}_\psi \mathbf{CoLearn}$ and $\mathbf{Min}_\psi \mathbf{FIN}$ differ for all Kolmogorov numberings; Theorem 3 shows that $\mathbf{Min}_\psi \mathbf{Ex}$ and $\mathbf{Min}_\psi \mathbf{CoLearn}$ differ for all Kolmogorov numberings.

Theorem 1. $(\forall \text{ Kolmogorov numbering } \psi)[\mathbf{Min}_\psi \mathbf{FIN} - \mathbf{FINITE} \neq \emptyset]$.

Proof. This is essentially the proof used by Freivalds [8] to show that for every Kolmogorov numbering ψ , there exists an infinite class of functions in $\mathbf{Min}_\psi \mathbf{Ex}$. We give the proof for completeness. Suppose a Kolmogorov numbering ψ is given. For $i > 0$, let h_i be defined as follows:

$$h_i(x) = \begin{cases} i & \text{if } x = 0; \\ 0 & \text{otherwise.} \end{cases}$$

Let c be such that,

$$(\forall j > 0)(\exists k \leq c * j)[\psi_k = h_j]$$

Since ψ is Kolmogorov numbering there exists such a c .

Let

$$\mathcal{S} = \{h_j \mid j > 0 \wedge \text{card}(\{i \mid i \leq c * j \wedge \psi_i(0) = j\}) \leq 2c\}.$$

It is easy to verify that \mathcal{S} is infinite.

Let $C_j = \{k \mid k \leq c * j \wedge \psi_k(0) = j\}$. Let p_j^l denote l th element enumerated in some standard, 1–1, effective in j , enumeration of C_j .

Now for $1 \leq i \leq 2c$, define \mathbf{M}^i as follows. \mathbf{M}^i on h_j behaves as follows: If p_j^i is defined, then \mathbf{M}^i on h_j outputs p_j^i as its only program; otherwise, \mathbf{M}^i does not output any program on h_j . Note that such \mathbf{M}^i can easily be constructed.

Now it is easy to verify that,

$$(\forall f \in \mathcal{S})(\exists l \mid 1 \leq l \leq 2c)[f \in \mathbf{Min}_\psi \mathbf{FIN}(\mathbf{M}^l)]$$

Since \mathcal{S} is infinite the theorem follows by Pigeon hole principle. \square

Definition 7. Suppose $\mathcal{C} \subseteq \mathcal{R}$ and $f \in \mathcal{C}$. Then f is said to be an accumulation point for \mathcal{C} iff $(\forall n)(\exists g \in \mathcal{C})[f \neq g \wedge (\forall x < n)[f(x) = g(x)]]$.

The following lemma is used in the proof of Theorem 2.

Lemma 1. Suppose $\mathcal{C} \subseteq \mathcal{R}$, and f is an accumulation point for \mathcal{C} . Then $\mathcal{C} \notin \mathbf{FIN}$.

Proof. Let \mathcal{C} and f be given as in the hypothesis. Suppose by way of contradiction $\mathbf{M} \mathbf{FIN}$ -identifies \mathcal{C} . Then there exists an $n \in \mathbb{N}$ such that $\mathbf{M}(f[n]) \neq ?$. Let g be such

that $g \neq f$ and $(\forall x \leq n)[f(x) = g(x)]$. Then \mathbf{M} fails to **FIN**-identify at least one of g and f . A contradiction. Thus no such \mathbf{M} can exist. \square

Theorem 2. $(\forall \text{ Kolmogorov numbering } \psi)[\mathbf{Min}_\psi \mathbf{CoLearn} - \mathbf{FIN} \neq \emptyset]$.

Proof. Suppose a Kolmogorov numbering ψ is given. Let h_0 be everywhere 0 function. For $i > 0$, define h_i as follows:

$$h_i(x) = \begin{cases} 1 & \text{if } x = i, \\ 0 & \text{otherwise.} \end{cases}$$

We will construct a class of functions $\mathcal{C} \in \mathbf{Min}_\psi \mathbf{CoLearn}$, such that $h_0 \in \mathcal{C}$ and for infinitely many k , $h_k \in \mathcal{C}$. Note that h_0 would thus be an accumulation point for \mathcal{C} . This, using Lemma 1, would imply the Theorem.

Let $z = \text{MinProg}_\psi(h_0)$.

Let c be such that $(\forall k > 0)[\text{MinProg}_\psi(h_k) \leq c * k]$ (note that since ψ is a Kolmogorov numbering, there exists such a c). For $k > 0$, let $C_k = \{j \mid j \leq c * k \wedge h_k[k + 1] \subseteq \psi_j\}$. Let p_k^l denote the l th program enumerated, if any, in some standard, 1–1, effective in k , enumeration of C_k .

Let c' be a constant > 2 . For $l \in \mathbb{N}^+$, consider machine \mathbf{M}^l , such that the following two properties are satisfied:

$$(1) \quad P_{\mathbf{M}^l}(A, h_0[y]) = \{i \mid i \leq y/c'\} - \{z\}.$$

(Note that this implies that $h_0 \in \mathbf{Min}_\psi \mathbf{CoLearn}(\mathbf{M}^l)$.)

$$(2) \quad P_{\mathbf{M}^l}(h_k[k + 1], h_k) = \begin{cases} N - \{p_k^l\} & \text{if } l \leq \text{card}(C_k), \\ \{z\} & \text{if } l > \text{card}(C_k). \end{cases}$$

Note that one can easily construct such \mathbf{M}^l .

Let

$$S = \{k \mid k > 0 \wedge \text{card}(C_k) \leq 2c\} - \{k \mid k > 0 \wedge \text{MinProg}_\psi(h_k) \leq k/c'\}$$

It is easy to verify that S is of infinite cardinality (since $c' > 2$). Furthermore, $(\forall k \in S)(\exists l \mid 1 \leq l \leq 2c)[h_k \in \mathbf{Min}_\psi \mathbf{CoLearn}(\mathbf{M}^l)]$. It follows, using Pigeon hole principle, that there exists an l , $1 \leq l \leq 2c$, such that $\mathbf{Min}_\psi \mathbf{CoLearn}(\mathbf{M}^l)$ contains an infinite subset of $\{h_k \mid k > 0\}$. Since $h_0 \in \mathbf{Min}_\psi \mathbf{CoLearn}(\mathbf{M}^l)$, it follows that there exists a $\mathcal{C} \in \mathbf{Min}_\psi \mathbf{CoLearn}$, such that $h_0 \in \mathcal{C}$, and for infinitely many k , $h_k \in \mathcal{C}$, as claimed. \square

Note that $\mathbf{FIN} - \mathbf{Min}_\psi \mathbf{CoLearn} \neq \emptyset$, for all acceptable numbering ψ . ($\mathcal{C} = \{f \mid (\forall x > 0)[f(x) = 0]\}$, witnesses the separation.)

As corollaries we have

Corollary 1. $(\forall \text{ Kolmogorov numbering } \psi)(\exists \mathcal{C})[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{CoLearn} \wedge (\forall \text{ Gödel numbering } \eta)[\mathcal{C} \notin \mathbf{Min}_\eta \mathbf{FIN}]]$.

Corollary 2. $(\forall \text{ Kolmogorov numbering } \psi)[\mathbf{Min}_\psi \mathbf{CoLearn} - \mathbf{Min}_\psi \mathbf{FIN} \neq \emptyset]$.

Theorem 3. $(\forall \text{ Kolmogorov numbering } \psi)(\exists \mathcal{C})[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex} \wedge (\forall \text{ Kolmogorov numbering } \psi')[\mathcal{C} \notin \mathbf{Min}_{\psi'} \mathbf{CoLearn}]]$.

Proof. Let β^n denote the n th Kolmogorov numbering (in some ordering of Kolmogorov numberings; note that we do not need the ordering to be effective – in fact any such ordering will not be effective).

We first define a computable numbering η (an appropriate subset of functions computed by the programs in the numbering η will serve as our diagonalizing class).

Intuitively, we consider the programs of η to be divided in different groups $G_i^j = \{p \mid l_i^j \leq p \leq u_i^j\}$, where $j \leq i$, and l_i^j, u_i^j are defined below. For each i , for some $s_i \leq i$, $G_i^{s_i}$ will provide us with a (large) set of functions \mathcal{S}_i , such that, for each $k, n \in N$, for large enough i , at most one of the functions in \mathcal{S}_i belongs to $\mathbf{Min}_{\beta^n} \mathbf{CoLearn}(\mathbf{M}_k)$. This will allow us to construct our diagonalizing class \mathcal{C} using techniques similar to that used in earlier theorems of this paper. We now proceed formally.

Let $l_0^0 = 0, u_0^0 = 0$.

For $i > 0$,

let $l_i^0 = u_{i-1}^{i-1} + 1$;

for $1 \leq j \leq i$, let $l_i^j = u_i^{j-1} + 1$.

For $i > 0, j \leq i$, let $u_i^j = (l_i^j + i^2) * 2$.

For $l_i^0 \leq r \leq u_i^i$, we define η_r according to the following staging construction (note that for each i , a different such staging construction, effective in i , is executed):

Let $\sigma_i^0 = \{(0, i)\}$. Go to stage 0.

Stage s

For $l_i^s \leq r \leq u_i^s$, let

$$\eta_r(x) = \begin{cases} \sigma_i^s(x) & \text{if } x < |\sigma_i^s|, \\ r & \text{if } x = |\sigma_i^s|, \\ 0 & \text{otherwise.} \end{cases}$$

Search for a r, n such that

$l_i^s \leq r \leq u_i^s$,

$n > |\sigma_i^s|$, and

$\text{card}(\{k \mid k < i \wedge \text{card}(\{x \mid x \leq i * u_i^i\} - P_{\mathbf{M}_k}(\Lambda, \eta_r[n])) \leq 1\}) \geq s + 1$.

(Note that the success of above search means that at least $s + 1$ of the machines in $\mathbf{M}_0, \mathbf{M}_1, \dots, \mathbf{M}_{i-1}$, output all but possibly one of the programs $\leq i * u_i^i$ on initial segments of $\eta_r[n]$.)

If and when such r, n are found, let $\sigma_i^{s+1} = \eta_r[n]$, and go to stage $s + 1$.

End Stage s

Note that for any i , the last stage executed is $\leq i$ (since, $\text{card}(\{k \mid k < i \wedge \text{card}(\{x \mid x \leq i * u_i^i\} - P_{\mathbf{M}_k}(\Lambda, \eta_r[n])) \leq 1\})$ is bounded by i).

For $i \in N$, let s_i denote the last stage executed in the above construction corresponding to i .

Claim 1. (a) Suppose $i \in N$, $j \leq s_i$, and $l_i^j \leq r \leq u_i^j$. Then $\eta_r \in \mathbf{ZEROSTAR}$ and $\eta_r(\max(\{x \mid \eta_r(x) \neq 0\})) = r$.

(b) Suppose $i \in N$, $s_i < j \leq i$, and $l_i^j \leq r \leq u_i^j$. Then η_r is everywhere undefined.

(c) If $r \neq r'$, then $\eta_r = \eta_{r'}$ implies η_r is everywhere undefined.

Proof. Parts (a) and (b) follow from definition of η_r and s_i . Part (c) follows from parts (a) and (b). \square

Let $\mathcal{S}_i = \{\eta_r \mid l_i^{s_i} \leq r \leq u_i^{s_i}\}$.

Let c_n be such that $(\forall r > 0)[\text{MinProg}_{\beta^n}(\eta_r) \leq c_n * r]$ (such c_n exist since β^n is a Kolmogorov numbering).

Let $E_{k,n}^i = \{f \mid f \in \mathcal{S}_i \wedge c_n \leq i \wedge f \in \mathbf{Min}_{\beta^n} \mathbf{CoLearn}(\mathbf{M}_k)\}$.

Claim 2. Let $k, n \in N$. $\mathbf{Min}_{\beta^n} \mathbf{CoLearn}(\mathbf{M}_k)$ contains at most finitely many functions in $[\bigcup_i \mathcal{S}_i] - [\bigcup_{\{i,k,n \mid k,n < i\}} E_{k,n}^i]$.

Proof. Follows from the definition of $E_{k,n}^i$. \square

Claim 3. Suppose $k, n < i$. Then $\text{card}(E_{k,n}^i) \leq 1$.

Proof. Suppose $k, n < i$, and $c_n \leq i$. Now for all $f \in \mathcal{S}_i$, $\text{MinProg}_{\beta^n}(f) \leq i * u_i^i$ (since $\text{MinProg}_{\eta_i}(f) \leq u_i^i$). Also for all $f \in \mathcal{S}_i$, $\sigma_i^{s_i} \subseteq f$.

Let $X = \{k' \mid k' < i \wedge \text{card}(\{x \mid x \leq i * u_i^i\} - P_{\mathbf{M}_{k'}}(\Lambda, \sigma_i^{s_i})) \leq 1\}$.

Now we consider two cases:

Case 1: $k \notin X$.

Note that in this case, we have that $\mathbf{M}_k(f)$ co-diverges on all $f \in \mathcal{S}_i$ (otherwise the search in stage s_i of the construction above would have succeeded). Thus $\text{card}(E_{k,n}^i) = 0$.

Case 2: $k \in X$.

In this case since,

$$\text{card}(\{x \mid x \leq i * u_i^i\} - P_{\mathbf{M}_k}(\Lambda, \sigma_i^{s_i})) \leq 1$$

we have that \mathbf{M}_k can $\mathbf{Min}_{\beta^n} \mathbf{CoLearn}$ at most one $f \in \mathcal{S}_i$. Thus $\text{card}(E_{k,n}^i) \leq 1$. \square

Now suppose ψ is a Kolmogorov numbering. Let c be such that $(\forall l > 0)[\text{MinProg}_{\psi}(\eta_l) \leq c * l]$. We will show that some machine $\mathbf{Min}_{\psi} \mathbf{Ex}$ -identifies an infinite subset of $[\bigcup_i \mathcal{S}_i] - [\bigcup_{\{i,k,n \mid k,n < i\}} E_{k,n}^i]$. This would prove the theorem (using Claim 2).

Let $\mathcal{S} = [\bigcup_i \mathcal{S}_i] - [[\bigcup_{\{i,k,n \mid k,n < i\}} E_{k,n}^i] \cup \{\eta_r \mid (\exists i)[l_i^{s_i} \leq r \leq u_i^{s_i}] \wedge \text{card}(\{p \mid p \leq c * r \wedge \eta_r[\sigma_i^{s_i}] + 1 \subseteq \psi_p\}) > 3c\}]$. It is easy to verify that \mathcal{S} is infinite (for $i > c$, at least $2u_i^{s_i}/3 - l_i^{s_i} - i^2$ of the functions in \mathcal{S}_i belong to \mathcal{S}).

For $l_i^{s_i} \leq r \leq u_i^{s_i}$, let $C_r = \{r' \mid r' \leq c * r \wedge \eta_r[|\sigma_i^{s_i}| + 1] \subseteq \psi_{r'}\}$. Let p_r^l denote the l th program, if any, enumerated in some standard, 1–1, effective in r , enumeration of C_r .

For $1 \leq l \leq 3c$, let \mathbf{M}^l be such that,

$$(\forall i)(\forall r \mid l_i^{s_i} \leq r \leq u_i^{s_i})[\mathbf{M}^l(\eta_r) = p_r^l].$$

Note that such \mathbf{M}^l can easily be constructed. It is easy to verify that, for each $\eta_r \in \mathcal{S}$, there exists an l , $1 \leq l \leq 3c$ such that $\eta_r \in \mathbf{Min}_\psi \mathbf{Ex}(\mathbf{M}^l)$. Thus, by pigeon hole principle, there exists an infinite subset of \mathcal{S} in $\mathbf{Min}_\psi \mathbf{Ex}$. Since $\mathcal{S} \subseteq [\bigcup_i \mathcal{S}_i] - [\bigcup_{\{i,k,n \mid k,n < i\}} E_{k,n}^i]$, we have that, an infinite subset of $[\bigcup_i \mathcal{S}_i] - [\bigcup_{\{i,k,n \mid k,n < i\}} E_{k,n}^i]$ belongs to $\mathbf{Min}_\psi \mathbf{Ex}$. The theorem follows using Claim 2. \square

A modification of the above proof can be used to show that

Theorem 4. $(\forall \text{ Kolmogorov numbering } \psi)(\exists \mathcal{C})[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex} \wedge (\forall \text{ Gödel numbering } \psi')[\mathcal{C} \notin \mathbf{Min}_{\psi'} \mathbf{CoLearn}]]$.

We omit the details. As a corollary to Theorem 3 we have,

Corollary 3. $(\forall \text{ Kolmogorov numbering } \psi)[\mathbf{Min}_\psi \mathbf{Ex} - \mathbf{Min}_\psi \mathbf{CoLearn} \neq \emptyset]$.

As a corollary from Theorem 1 and Corollaries 2 and 3 we have

Corollary 4. $(\forall \text{ Kolmogorov numbering } \psi)[\mathbf{FINITE} \subset \mathbf{Min}_\psi \mathbf{FIN} \subset \mathbf{Min}_\psi \mathbf{CoLearn} \subset \mathbf{Min}_\psi \mathbf{Ex}]$.

4.2. Recursively enumerable classes and minimal identification in Kolmogorov/Gödel numberings

The next three theorems consider the question about whether recursively enumerable classes can be minimally identified in Gödel or Kolmogorov numberings.

Theorem 5. $(\exists \text{ infinite r.e. } \mathcal{C})(\exists \text{ Kolmogorov numbering } \psi)[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex}]$.

Proof. It was shown in [8] that there exists a Kolmogorov numbering ψ such that $\{f \mid (\forall x)[f(x) = f(0)]\} \in \mathbf{Min}_\psi \mathbf{Ex}$. In fact it can be shown that for all $\mathcal{C} \in \mathbf{FIN}$, there exists a Kolmogorov numbering ψ , such that $\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex}$. \square

Theorem 6. $(\forall \text{ infinite r.e. } \mathcal{C})(\forall \text{ Gödel numbering } \psi)[\mathcal{C} \notin \mathbf{Min}_\psi \mathbf{CoLearn}]$.

Proof. Suppose by way of contradiction that \mathbf{M} , an r.e. infinite class \mathcal{C} , and a Gödel numbering ψ , are such that $\mathcal{C} \subseteq \mathbf{Min}_\psi \mathbf{CoLearn}(\mathbf{M})$.

It follows that, for all c , there exists an $f \in \mathcal{C}$ such that $P_{\mathbf{M}}(\mathcal{A}, f) \supseteq \{x \mid x \leq c\}$. Note that, since \mathcal{C} is r.e., one can search, effectively in c , for such a function f .

Thus, by implicit use of Kleene's recursion theorem [14], there exists an e , such that $\psi_e \in \mathcal{C}$, and $P_M(A, \psi_e) \supseteq \{x \mid x \leq e\}$. But then $\psi_e \notin \mathbf{Min}_\psi \mathbf{CoLearn}(M)$. A contradiction. Thus $(\forall$ infinite r.e. $\mathcal{C})(\forall$ Gödel numbering $\psi)[\mathcal{C} \not\subseteq \mathbf{Min}_\psi \mathbf{CoLearn}]$. \square

Theorem 7. $(\exists$ Kolmogorov numbering $\psi)(\exists$ infinite co-r.e. $L)[\{\psi_i \mid i \in L\} \in \mathbf{Min}_\psi \mathbf{FIN}]$.

Proof. Let f_k be defined as follows:

$$f_k(x) = \begin{cases} k & \text{if } x = 0, \\ 0 & \text{otherwise.} \end{cases}$$

Without loss of generality suppose that φ is a Kolmogorov numbering.

Let ψ be defined as follows:

$$\psi_j = \begin{cases} \varphi_i & \text{if } j = 3i, \\ f_j & \text{if } j \text{ is not divisible by 3.} \end{cases}$$

Let

$$\bar{L} = \{3i \mid i \in N\} \cup \{j \mid j \text{ is not divisible by 3} \wedge (\exists i < j/3)[\varphi_i(0) = j]\}.$$

Note that \bar{L} is r.e. and coinfinite. Consider M , which on f_k , outputs k as its only program. It is easy to verify that $\mathcal{C} = \{\psi_j \mid j \in L\} \subseteq \mathbf{Min}_\psi \mathbf{FIN}(M)$. \square

4.3. Minimal identification in Gödel numberings vs. Kolmogorov numberings

In this section we compare the effects of considering Gödel numberings versus Kolmogorov numbering on minimal identification. Specifically, we show that, for each of the three identification types, **FIN**, **CoLearn**, **Ex**, discussed in this paper, there exists a class of functions, \mathcal{C} , which can be identified using minimal programs in some Gödel numbering but cannot be identified using minimal programs in any Kolmogorov numbering.

Theorem 8. $(\exists$ Gödel numbering $\psi)(\exists \mathcal{C})[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{FIN} \wedge (\forall$ Kolmogorov numbering $\psi')[\mathcal{C} \not\subseteq \mathbf{Min}_{\psi'} \mathbf{CoLearn}]]$.

Proof. Let h_i be a function defined as follows:

$$h_i(x) = \begin{cases} i & \text{if } x = 0, \\ 0 & \text{otherwise.} \end{cases}$$

Let ψ_i be defined as follows:

$$\psi_i = \begin{cases} \varphi_l & \text{if } i = l^4, \\ h_i & \text{if for all } l, i \neq l^4. \end{cases}$$

Clearly, ψ is a Gödel numbering. Consider M' , which on h_i , outputs i as its only program. Let $\mathcal{C} = \{h_i \mid \text{MinProg}_\psi(h_i) = i\}$. Clearly, $\mathcal{C} \subseteq \mathbf{Min}_\psi \mathbf{FIN}(M')$.

Claim 4. *At most $r + 1$ of the functions in $\{h_i \mid r^4 < i < (r + 1)^4\}$ are not in \mathcal{C} .*

Proof. Note that $\text{card}(\{i \mid i < (r + 1)^4 \wedge \psi_i \neq h_i\}) \leq r + 1$. Thus since h_i 's are distinct, $\text{card}(\{i \mid r^4 < i < (r + 1)^4 \wedge \text{MinProg}_\psi(h_i) \neq i\}) \leq r + 1$. The claim follows. \square

Suppose \mathbf{M} and a Kolmogorov numbering ψ' is given. We will show that $\mathcal{C} \not\subseteq \text{Min}_{\psi'}\text{CoLearn}(\mathbf{M})$. To show this we, construct a recursive function q (using operator recursion theorem [3]) such that, for some i , $\varphi_{q(i)}$ has a small enough program in ψ' , but \mathbf{M} on $\varphi_{q(i)}$ does not co-converge to a small enough program.

By Operator recursion theorem [3], there exists a recursive function q , such that (partial) functions $\varphi_{q(\cdot)}$ may be defined as follows.

Let $l_0 = 0$. l_s is used to denote the least l such that $\varphi_{q(l)}$ has not been defined on any input before stage s . Go to stage 0.

Begin stage s

1. Search for an $r > l_s$, $S \subseteq \{x \mid r^4 < x < (r + 1)^4\}$, such that $\text{card}(S) = r + 2$ and, for all $i \in S$, $\{x \mid x \leq (l_s + r + 2)^2\} \subseteq P_{\mathbf{M}}(\lambda, h_i)$.
2. Let S be as found in step 1. Let $\varphi_{q(l_s+x)}$, $x \leq r + 1$, be the $r + 2$ functions in $\{h_i \mid i \in S\}$.
3. Let $l_{s+1} = l_s + r + 2$.

End stage s

Now we consider the following cases.

Case 1: Stage s starts but does not halt.

Non-success of the search at step 1, implies that, for any $r > l_s$, \mathbf{M} can co-learn (in numbering ψ) at most $(l_s + r + 2)^2 + r + 1 \leq (2r + 2)^2 + r + 1$ of the functions in $\{h_i \mid r^4 < i < (r + 1)^4\}$. Now, for each r , by Claim 4, at least $(r + 1)^4 - r^4 - 1 - (r + 1)$ of the functions in $\{h_i \mid r^4 < i < (r + 1)^4\}$, are in \mathcal{C} .

Thus, since $(r + 1)^4 - r^4 - 1 - (r + 1) > (2r + 2)^2 + r + 1$, for large enough r , non-success of step 1 in stage s implies $\mathcal{C} \not\subseteq \text{Min}_{\psi'}\text{CoLearn}(\mathbf{M})$.

Case 2: All stages halt.

In this case note that, by Claim 4, for each stage, at least one of the $r + 2$ functions found in step 1 is in \mathcal{C} . It follows that $(\exists^\infty i)[\varphi_{q(i)} \in \mathcal{C} \wedge \mathbf{M}$ on $\varphi_{q(i)}$ does not co-converge to a program $\leq i^2]$. But since, for some constant c , for all i , $\text{MinProg}_{\psi'}(\varphi_{q(i)}) \leq c * i$, we have

$$(\exists^\infty i)[\varphi_{q(i)} \in (\mathcal{C} - \text{Min}_{\psi'}\text{CoLearn}(\mathbf{M}))].$$

Thus $\mathcal{C} \not\subseteq \text{Min}_{\psi'}\text{CoLearn}(\mathbf{M})$. \square

As corollaries we have,

Corollary 5. $(\exists \mathcal{C})(\exists \text{ Gödel numbering } \psi)[\mathcal{C} \in \text{Min}_\psi\text{FIN} \wedge (\forall \text{ Kolmogorov numbering } \eta)[\mathcal{C} \notin \text{Min}_\eta\text{FIN}]]$.

Corollary 6. $(\exists \mathcal{C})(\exists \text{ Gödel numbering } \psi)[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{CoLearn} \wedge (\forall \text{ Kolmogorov numbering } \eta)[\mathcal{C} \notin \mathbf{Min}_\eta \mathbf{CoLearn}]]$.

Note that we cannot strengthen the above theorem to show, $(\exists \text{ Gödel numbering } \psi)(\exists \mathcal{C})[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{FIN} \wedge (\forall \text{ Kolmogorov numbering } \psi')[\mathcal{C} \notin \mathbf{Min}_{\psi'} \mathbf{Ex}]]$. This is so because, $(\forall \mathcal{C} \in \mathbf{FIN})[(\exists \text{ Kolmogorov numbering } \psi')[\mathcal{C} \in \mathbf{Min}_{\psi'} \mathbf{Ex}]]$. We do not know at this point whether, $(\exists \mathcal{C})(\exists \text{ Gödel numbering } \psi)[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{CoLearn} \wedge (\forall \text{ Kolmogorov numbering } \eta)[\mathcal{C} \notin \mathbf{Min}_\eta \mathbf{Ex}]]$. However,

Theorem 9. $(\exists \mathcal{C})(\exists \text{ Gödel numbering } \psi)[\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex} \wedge (\forall \text{ Kolmogorov numbering } \eta)[\mathcal{C} \notin \mathbf{Min}_\eta \mathbf{Ex}]]$.

Proof. Let F be a partial recursive function, such that $F(k, i, x)$, denotes the output of the i th program in the k th numbering on input x . Note that $F(k, \cdot, \cdot)$ is a numbering, and $F(k, i, \cdot)$ denotes the function computed by the i th program in this numbering.

We will construct a recursive function $g(k, i, j)$ using parameterized recursion theorem. We will have that, for all k, i, j , $\text{zeroext}(\varphi_{g(k, i, j)}) \in \mathbf{ZEROSTAR}$.

Intuitively our plan is as follows:

(A) We try to make $\text{zeroext}(\varphi_{g(k, i, j)}) \notin \mathbf{Min}_{F(k, \cdot, \cdot)} \mathbf{Ex}(\mathbf{M}_i)$. For this we use a technique used by Chen [6] to show that $\mathbf{ZEROSTAR} \not\subseteq \mathbf{MEx}$ (we refer the reader to Chen [6] for definition of \mathbf{MEx} -identification). However, since we do not know the reduction from $\varphi_{g(k, i, \cdot)}$ to the Kolmogorov numbering $F(k, \cdot, \cdot)$, we may not always be successful. We will however be successful for large enough j (where large enough depends only on i and k).

(B) Using g , we will construct a class \mathcal{C} (which contains, for each i, k , infinitely many of $\text{zeroext}(\varphi_{g(k, i, \cdot)})$) and a Gödel numbering ψ such that $\mathcal{C} \in \mathbf{Min}_\psi \mathbf{Ex}$. For this purpose we need to code $\langle k, i, j \rangle$ in $\varphi_{g(k, i, j)}$.

This would prove the theorem. We now proceed formally.

By parameterized s-m-n theorem [14] there exists a recursive 1–1 function g such that $\varphi_{g(k, i, j)}$, may be defined as follows.

Begin definition of $\varphi_{g(k, i, j)}$

Let $\text{Cancel} = \emptyset$.

Let $\varphi_{g(k, i, j)}(0) = \langle k, i, j \rangle$.

Let x_s denote the least x such that $\varphi_{g(k, i, j)}(x)$ has not been defined before stage s .

Go to stage 0.

Begin stage s

0. Dovetail steps 1 and 2, until one of them succeeds. If step 1 succeeds (before step 2, if ever) then go to step 3. If step 2 succeeds (before step 1, if ever) then go to step 4.

1. Search for an $l < j^2$, such that $l \notin \text{Cancel}$ and $F(k, l, x_s) \downarrow$.

2. Suppose $f = \text{zeroext}(\varphi_{g(k, i, j)}[x_s])$. Search for $m > x_s$, such that $\mathbf{M}_i(f[m]) \geq j^2$.

3. Let l be as found in step 1. Let $\varphi_{g(k,i,j)}(x_s) = F(k, l, x_s) + 1$.

Let $\text{Cancel} = \text{Cancel} \cup \{l\}$.

Go to stage $s + 1$.

4. Let m be as found in step 2. For $x_s \leq x \leq m$, let $\varphi_{g(k,i,j)}(x) = 0$.

Go to stage $s + 1$.

End stage s

End definition of $\varphi_{g(k,i,j)}$

Claim 5. Suppose $F(k, \cdot, \cdot)$ is a Kolmogorov numbering. Then for all i , for all but finitely many j , $\text{zeroext}(\varphi_{g(k,i,j)}) \notin \text{Min}_{F(k, \cdot, \cdot)} \text{Ex}(\mathbf{M}_i)$.

Proof. Fix \mathbf{M}_i . Suppose $F(k, \cdot, \cdot)$ is a Kolmogorov numbering. Then for all but finitely many j , $\text{MinProg}_{F(k, \cdot, \cdot)}(\varphi_{g(k,i,j)}) < j^2$.

Suppose j is such that $\text{MinProg}_{F(k, \cdot, \cdot)}(\varphi_{g(k,i,j)}) < j^2$. Let $f = \text{zeroext}(\varphi_{g(k,i,j)})$. We will show that $f \notin \text{Min}_{F(k, \cdot, \cdot)} \text{Ex}(\mathbf{M}_i)$. Consider the following two cases.

Case 1: $(\exists n)[\mathbf{M}_i(f[n]) \geq j^2]$.

In this case, due to the success of step 2 in the construction of $\varphi_{g(k,i,j)}$ infinitely often, we have, $f = \varphi_{g(k,i,j)} \notin \text{Min}_{F(k, \cdot, \cdot)} \text{Ex}(\mathbf{M}_i)$.

Case 2: $(\forall n)[\mathbf{M}_i(f[n]) < j^2]$.

By the construction of $\varphi_{g(k,i,j)}$, it follows that $(\forall l < j^2)[l \in \text{Cancel} \vee F(k, l, \cdot)$ is not total]. Therefore, $(\forall l < j^2)[F(k, l, \cdot) \neq f]$. Thus $f \notin \text{Min}_{F(k, \cdot, \cdot)} \text{Ex}(\mathbf{M}_i)$. \square

Define $\mathcal{C}_{\langle k,i,j \rangle}$ as follows:

$$\mathcal{C}_{\langle k,i,j \rangle} = \{\text{zeroext}(\varphi_{g(k,i,j)}[x]) \mid x = 1 \vee \varphi_{g(k,i,j)}(x-1) \downarrow \neq 0\}.$$

Note that $\text{zeroext}(\varphi_{g(k,i,j)}) \in \mathcal{C}_{\langle k,i,j \rangle}$, and $\text{card}(\mathcal{C}_{\langle k,i,j \rangle}) \leq j^2 + 1$ (since step 1 can succeed only j^2 times). Moreover the functions in $\mathcal{C}_{\langle k,i,j \rangle}$ are 1-1 enumerable effectively in $\langle k, i, j \rangle$.

Let

$$S = \{\langle k, i, j \rangle \mid (\forall p \leq \sqrt{j})[\varphi_p(0) \neq \langle k, i, j \rangle]\}.$$

Let

$$\mathcal{C} = \bigcup_{\langle k,i,j \rangle \in S} \mathcal{C}_{\langle k,i,j \rangle}.$$

Note that for each k, i , there exist infinitely many j , such that $\mathcal{C}_{\langle k,i,j \rangle} \subseteq \mathcal{C}$. Thus, for all i, k , there exists infinitely many j such that $\text{zeroext}(\varphi_{g(k,i,j)}) \in \mathcal{C}$. It follows from Claim 5 that, $(\forall \text{ Kolmogorov numbering } \psi')[\mathcal{C} \notin \text{Min}_{\psi'} \text{Ex}]$.

We now construct a Gödel numbering ψ , such that $\mathcal{C} \in \text{Min}_{\psi} \text{Ex}$. Let $\text{gap}(l) = l^4 + 2$. Let $h(0) = 0$; for $l \in \mathbb{N}$, let $h(l+1) = 1 + h(l) + \text{gap}(l)$.

Let $\psi_{h(l+1)} = \varphi_l$. Note that this makes ψ a Gödel numbering.

For $h(l) < x < h(l+1)$, ψ_x is defined as follows. If $\varphi_l(0) \downarrow = \langle k, i, j \rangle$ and $\text{card}(\mathcal{C}_{\langle k, i, j \rangle}) \geq x - h(l)$, then let $\psi_x = (x - h(l))$ -th function in $\mathcal{C}_{\langle k, i, j \rangle}$ (in some standard, 1–1, effective in k, i, j , enumeration of $\mathcal{C}_{\langle k, i, j \rangle}$). Otherwise let ψ_x be the everywhere undefined function.

Claim 6. $\mathcal{C} \in \text{Min}_\psi \text{Ex}$.

Proof. Suppose, $\langle k, i, j \rangle \in S$. Suppose l is such that $\varphi_l(0) = \langle k, i, j \rangle$. Then we have $l > \sqrt{j}$. Thus, $j^2 + 1 \leq l^4 + 1$. Hence, $\mathcal{C}_{\langle k, i, j \rangle} \subseteq \{\psi_r \mid h(l) < r < h(l+1)\}$.

Thus for all $f \in \mathcal{C}$, we have:

$$h(l) < \text{MinProg}_\psi(f) < h(l+1)$$

where $l = \min(\{r \mid \varphi_r(0) = f(0)\})$.

Thus, in particular, (A) for all $f \in \mathcal{C}$, $\text{MinProg}_\psi(f)$ is not of the form $h(l)$, for any l . Moreover the construction of ψ gives us the following: (B) if $h(l) < x < h(l+1)$, then either ψ_x is total or ψ_x is everywhere undefined.

Using properties (A) and (B) of ψ , it is easy to show that $\mathcal{C} \in \text{Min}_\psi \text{Ex}$. \square

Recall that in every Kolmogorov numbering ψ , $\text{Min}_\psi \text{FIN}$, $\text{Min}_\psi \text{CoLearn}$, $\text{Min}_\psi \text{Ex}$ are separated. However, as the following theorem shows, in Gödel numbering this may not be the case.

Theorem 10. For all $\alpha_1, \alpha_2, \alpha_3 \in \{=, \subset\}$, it is possible to construct a Gödel Numbering η such that $\text{FINITE} \alpha_1 \text{Min}_\eta \text{FIN} \alpha_2 \text{Min}_\eta \text{CoLearn} \alpha_3 \text{Min}_\eta \text{Ex}$.

Proof. The idea is essentially to interleave the diagonalizations for the relevant proper subset construction with the Gödel numbering in which no infinite set of functions is MinEx -identifiable.

Lemmas 2–4 below give us (non-universal) computable numberings ψ^1, ψ^2, ψ^3 , and monotone increasing recursive functions g^1, g^2, g^3 such that properties (A) to (G) below are satisfied.

In the following, let $\mathcal{C}^1 = \{\psi_j^1 \mid \psi_j^1 \in \mathcal{R}\}$, $\mathcal{C}^2 = \{\psi_j^2 \mid \psi_j^2 \in \mathcal{R}\}$, and $\mathcal{C}^3 = \{\psi_j^3 \mid \psi_j^3 \in \mathcal{R}\}$.

(A) $\mathcal{C}^1, \mathcal{C}^2, \mathcal{C}^3$ are infinite and pairwise disjoint.

(B) $\mathcal{C}^1 \in \text{Min}_{\psi^1} \text{FIN}$.

(C) $\mathcal{C}^2 \in \text{Min}_{\psi^2} \text{CoLearn}$.

(D) $\mathcal{C}^3 \in \text{Min}_{\psi^3} \text{Ex}$.

(E) No infinite subset of \mathcal{C}^2 belongs to $\text{Min}_{\psi^2} \text{FIN}$.

(F) No infinite subset of \mathcal{C}^3 belongs to $\text{Min}_{\psi^3} \text{CoLearn}$.

(G) For each $i \in \{1, 2, 3\}$, there exist infinitely many $j \in N$ such that, $\text{card}(\{\psi_r^i \mid r \leq g^i(j) \wedge \psi_r^i \in \mathcal{C}^i\}) > 2j$.

Using the above numberings, we construct a Gödel numbering η satisfying the theorem.

Suppose β is the Gödel numbering in which no infinite class of functions can be $\text{Min}_\beta \text{Ex}$ -identified. Intuitively we would like to interleave the numberings $\beta, \psi^1, \psi^2, \psi^3$,

so that, for $i \in \{1, 2, 3\}$, $g^i(j)$ -th program in ψ^i appears before j th program in β in the interleaving.

For this purpose let H be a 1–1 function from $\{(x, y) \mid 1 \leq x \leq 4 \wedge y \in N\}$ to N , such that the following two properties are satisfied.

(1) For $i \in \{1, 2, 3\}$, $H(i, g^i(j)) < H(4, j)$.

(2) For each $i \in \{1, 2, 3, 4\}$, $H(i, j)$ is a monotone increasing function of j .

Note that such a function H can be easily constructed. For $i \in \{1, 2, 3, 4\}$ and $k \in N$, let $\eta_{H(i, k)}$ be defined as follows.

$$\eta_{H(i, k)}(x) = \begin{cases} \beta_k(x) & \text{if } i = 4, \\ \psi_k^i(x) & \text{if } i \in \{1, 2, 3\} \text{ and } \alpha_i \text{ is } \subset, \\ \uparrow & \text{if } i \in \{1, 2, 3\} \text{ and } \alpha_i \text{ is } =. \end{cases}$$

Claim 7. (a) If $\mathcal{C} \subseteq \mathbf{Min}_\eta \mathbf{FIN}(\mathbf{M})$, then for all but finitely many $f \in \mathcal{C}$, $\mathbf{M}(f) \in \{H(i, k) \mid k \in N \wedge i = 1\}$.

(b) If $\mathcal{C} \subseteq \mathbf{Min}_\eta \mathbf{CoLearn}(\mathbf{M})$, then for all but finitely many $f \in \mathcal{C}$, $\mathbf{M}(f)$ co-converges to a member of $\{H(i, k) \mid k \in N \wedge i \in \{1, 2\}\}$.

(c) If $\mathcal{C} \subseteq \mathbf{Min}_\eta \mathbf{Ex}(\mathbf{M})$, then for all but finitely many $f \in \mathcal{C}$, $\mathbf{M}(f) \in \{H(i, k) \mid k \in N \wedge i \in \{1, 2, 3\}\}$.

Proof. Note that if, $H(i, k) = \text{MinProg}_\eta(f)$, then $k = \text{MinProg}_{\psi^i}(f)$, where we let $\psi^4 = \beta$. Thus, the claim follows from the construction of η and the facts that

- (i) no infinite subset of \mathcal{R} belongs to $\mathbf{Min}_\beta \mathbf{Ex}$,
- (ii) no infinite subset of \mathcal{C}^3 belongs to $\mathbf{Min}_{\psi^3} \mathbf{CoLearn}$, and
- (iii) no infinite subset of \mathcal{C}^2 belongs to $\mathbf{Min}_{\psi^2} \mathbf{FIN}$. \square

Claim 8. The following hold.

(a) Suppose α_1 is \subset . Then $\{\psi_i^1 \mid \psi_i^1 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^1) = H(1, i)\} \in \mathbf{Min}_\eta \mathbf{FIN}$. Moreover, $\{\psi_i^1 \mid \psi_i^1 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^1) = H(1, i)\}$ is infinite.

(b) Suppose α_2 is \subset . Then $\{\psi_i^2 \mid \psi_i^2 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^2) = H(2, i)\} \in \mathbf{Min}_\eta \mathbf{CoLearn}$. Moreover, $\{\psi_i^2 \mid \psi_i^2 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^2) = H(2, i)\}$ is infinite.

(c) Suppose α_3 is \subset . Then $\{\psi_i^3 \mid \psi_i^3 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^3) = H(3, i)\} \in \mathbf{Min}_\eta \mathbf{Ex}$. Moreover, $\{\psi_i^3 \mid \psi_i^3 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^3) = H(3, i)\}$ is infinite.

Proof. We show part (a). Proof of other parts are similar. Suppose α_1 is \subset . Suppose \mathbf{M} is such that $\mathcal{C}^1 \subseteq \mathbf{Min}_{\psi^1} \mathbf{FIN}(\mathbf{M})$. Let \mathbf{M}' be defined as follows. $\mathbf{M}'(\sigma) = H(1, \mathbf{M}(\sigma))$. Clearly, $\{\psi_i^1 \mid \psi_i^1 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^1) = H(1, i)\} \in \mathbf{Min}_{\psi^1} \mathbf{FIN}(\mathbf{M}')$.

Now, since $\mathcal{C}^1, \mathcal{C}^2, \mathcal{C}^3$ are pairwise disjoint, we have, for infinitely many j , $\text{card}(\{i \mid i \leq g^1(j) \wedge \psi_i^1 \in \mathcal{C}^1 \wedge H(1, i) = \text{MinProg}_\eta(\psi_i^1)\}) \geq 2j + 1 - j$.

It follows that $\{\psi_i^1 \mid \psi_i^1 \in \mathcal{R} \wedge \text{MinProg}_\eta(\psi_i^1) = H(1, i)\}$ is infinite. \square

Theorem 10 follows from the above two claims. \square

Lemma 2. *There exists (non-universal) computable numbering ψ^1 and monotone increasing recursive function g^1 , which satisfy properties (A) to (C) below.*

Below let $\mathcal{C}^1 = \{\psi_j^1 \mid \psi_j^1 \in \mathcal{R}\}$.

(A) For all $f \in \mathcal{C}^1$, $f(0) = 1$.

(B) $\mathcal{C}^1 \in \mathbf{Min}_{\psi^1} \mathbf{FIN}$.

(C) There exists infinitely many $j \in N$ such that, $\text{card}(\{\psi_r^1 \mid r \leq g^1(j) \wedge \psi_r^1 \in \mathcal{C}^1\}) > 2j$.

Proof. Let

$$\psi_k^1(x) = \begin{cases} 1 & \text{if } x = 0, \\ k & \text{otherwise.} \end{cases}$$

Let $g^1(j) = 2j + 2$. It is easy to verify that the properties (A) to (C) are satisfied. \square

Lemma 3. *There exists (non-universal) computable numbering ψ^2 and monotone increasing recursive function g^2 , which satisfy properties (A) to (D) below.*

Below let $\mathcal{C}^2 = \{\psi_j^2 \mid \psi_j^2 \in \mathcal{R}\}$.

(A) For all $f \in \mathcal{C}^2$, $f(0) = 2$.

(B) $\mathcal{C}^2 \in \mathbf{Min}_{\psi^2} \mathbf{CoLearn}$.

(C) No infinite subset of \mathcal{C}^2 belongs to $\mathbf{Min}_{\psi^2} \mathbf{FIN}$.

(D) There exists infinitely many $j \in N$ such that, $\text{card}(\{\psi_r^2 \mid r \leq g^2(j) \wedge \psi_r^2 \in \mathcal{C}^2\}) > 2j$.

Proof. Let $\sigma_0, \sigma_1, \dots$ be an 1–1 enumeration of all elements of INIT. We assume without loss of generality that $|\sigma_i| \leq i$. Let X be a recursive function from N^2 to N such that the following properties are satisfied (note that such an X can easily be constructed).

(1) $(\forall j)[\sigma_{X(j,0)} = \{(0,2), (1,j)\}]$.

(2) $(\forall j, l)[\sigma_{X(j,l)} \subseteq \sigma_{X(j,l+1)}]$.

(3) $(\forall j)[\lim_{l \rightarrow \infty} X(j, l) \downarrow]$.

(4) $(\forall j, l)[X(j, l) \neq X(j, l+1) \Rightarrow |\sigma_{X(j,l+1)}| < l]$.

(5) For any given j , suppose $\tau_j = \sigma_{\lim_{l \rightarrow \infty} X(j, l)}$. Then $(\forall k < j)[(\exists \tau' \supseteq \tau_j)[\mathbf{M}_k(\tau') \neq ?] \Rightarrow [\mathbf{M}_k(\tau_j) \neq ?]]$.

Intuitively, τ_j above denotes a sequence such that all \mathbf{M}_k , $k < j$, which output program on some extension of τ_j , output a program on τ_j . Conditions, (1)–(4) above just impose some restrictions on the search of such τ_j , which is used for implementing the diagonalization.

For $j \in N$, let l_j be the least value of l such that $(\forall l' > l)[X(j, l) = X(j, l')]$ (note that since $\lim_{l \rightarrow \infty} X(j, l) \downarrow$, l_j is well defined). Intuitively, l_j is just the convergence point for $X(j, \cdot)$.

Define h as follows:

$$h(0) = 0. \quad h(k+1) = h(k) + 3k + 2.$$

For $r \in N$, $r < 3 * \langle j, l \rangle + 2$, let

$$f_{h(\langle j, l \rangle) + r}(x) = \begin{cases} \sigma_{X(j, l)}(x) & \text{if } x < |\sigma_{X(j, l)}|, \\ h(\langle j, l \rangle) + r + 1 & \text{if } x = |\sigma_{X(j, l)}|, \\ 0 & \text{otherwise.} \end{cases}$$

Now, for $r < 3 * \langle j, l \rangle + 2$, let $\psi_{h(\langle j, l \rangle) + r}^2$ be defined so that the following two properties are satisfied:

$$(a) \quad \psi_{h(\langle j, l \rangle) + r}^2 \subseteq f_{h(\langle j, l \rangle) + r};$$

(b) $\psi_{h(\langle j, l \rangle) + r}^2 = f_{h(\langle j, l \rangle) + r}$ iff $[l = l_j \wedge (\forall m < j)[\mathbf{M}_m \text{ on } f_{h(\langle j, l \rangle) + r} \text{ does not finitely converge to } h(\langle j, l \rangle) + r]]$.

Intuitively, aim of part (b) is to make $\psi_{h(\langle j, l \rangle) + r}^2$ total iff the convergence point of $X(j, \cdot)$ is l , and no machine \mathbf{M}_m , $m < j$, finitely converges on $f_{h(\langle j, l \rangle) + r}$ to $h(\langle j, l \rangle) + r$.

Let $g^2(k) = h(k+1)$. We now show that ψ^2 and g^2 satisfy the conditions for the lemma.

Claim 9. No machine $\text{Min}_{\psi^2}\text{FIN}$ -identifies an infinite subset of $\mathcal{C}^2 = \{\psi_i^2 \mid \psi_i^2 \in \mathcal{R}\}$.

Proof. Clearly, by definition of $\psi_{h(\langle j, l \rangle) + r}^2$, where $r < 3 * \langle j, l \rangle + 2$, we have

$$(1) \quad \psi_{h(\langle j, l \rangle) + r}^2 \in \mathcal{R} \Rightarrow l = l_j.$$

(2) For all j , for all $m < j$, \mathbf{M}_m does not $\text{Min}_{\psi^2}\text{Fin}$ -identify any function in $\{\psi_{h(\langle j, l \rangle) + r}^2 \mid r < 3 * \langle j, l \rangle + 2 \wedge \psi_{h(\langle j, l \rangle) + r}^2 \in \mathcal{R}\}$.

The claim follows from above. \square

Claim 10. Let $\mathcal{C}^2 = \{\psi_i^2 \mid \psi_i^2 \in \mathcal{R}\}$. Then, for infinitely many $k \in N$, $\text{card}(\{\psi_r^2 \mid r \leq g^2(k) \wedge \psi_r^2 \in \mathcal{C}^2\}) > 2k$. Moreover, $\mathcal{C}^2 \in \text{Min}_{\psi^2}\text{CoLearn}$.

Proof. For all j , for $r < 3 * \langle j, l_j \rangle + 2$, $\sigma_{X(j, l_j)} \subseteq f_{h(\langle j, l_j \rangle) + r}$. Moreover, by property (5) of X , for all $m < j$, if \mathbf{M}_m outputs a program on $f_{h(\langle j, l_j \rangle) + r}$, then \mathbf{M}_m outputs a program on $\sigma_{X(j, l_j)}$. It thus follows from the construction of ψ^2 that, there exist at least $3 * \langle j, l_j \rangle + 2 - j$ distinct values for $r < 3 * \langle j, l_j \rangle + 2$, such that $\psi_{h(\langle j, l_j \rangle) + r}^2 = f_{h(\langle j, l_j \rangle) + r} \in \mathcal{R}$. Now, since f_i 's are distinct, it follows that, for all j ,

$$\text{card}(\{\psi_r^2 \mid r \leq g^2(\langle j, l_j \rangle) \wedge \psi_r^2 \in \mathcal{C}^2\}) \geq 3 * \langle j, l_j \rangle + 2 - j > 2 * \langle j, l_j \rangle$$

(recall that according to the assumption on our pairing function, $\langle i, j \rangle \geq \max(\{i, j\})$).

We now give a machine \mathbf{M} such that $\mathcal{C}^2 \subseteq \text{Min}_{\psi^2}\text{CoLearn}(\mathbf{M})$. For this it is sufficient to construct a machine \mathbf{M} such that \mathbf{M} on f_k co-converges to k .

First note that $l > |\sigma_{X(j, l)}| - 3$ (we need -3 just to address the case of $l = 0$). This implies (from the definition of $f_{h(\langle j, l \rangle) + r}$) that, for all j , for all $r < 3 * \langle j, l \rangle + 2$,

$$h(\langle j, l \rangle) + r \geq \langle j, l \rangle \geq l \geq |\sigma_{X(j, l)}| - 3 \geq \max(\{x \mid f_{h(\langle j, l \rangle) + r}(x) \neq 0\}) - 4.$$

Also note that, for all k , $f_k(\max(\{x \mid f_k(x) \neq 0\})) = k + 1$. Let \mathbf{M} be such that:

$$\mathbf{M}(\sigma) = \begin{cases} ? & \text{if } |\sigma| \leq 5, \\ \min(N - \{\sigma(\max(\{x \mid \sigma(x) \neq 0\})) - 1\}) & \text{otherwise.} \end{cases}$$

From the discussion above, it is easy to see that \mathbf{M} on f_k co-converges to k . Thus $\mathcal{C}^2 \subseteq \mathbf{Min}_{\psi^2} \mathbf{CoLearn}(\mathbf{M})$. \square

Lemma 3 follows from the above two claims. \square

Lemma 4. *There exists (non-universal) computable numbering ψ^3 and monotone increasing recursive function g^3 , which satisfy properties (A) to (D) below.*

Below let $\mathcal{C}^3 = \{\psi_j^3 \mid \psi_j^3 \in \mathcal{R}\}$.

(A) For all $f \in \mathcal{C}^3$, $f(0) = 3$.

(B) $\mathcal{C}^3 \in \mathbf{Min}_{\psi^3} \mathbf{Ex}$.

(C) No infinite subset of \mathcal{C}^3 belongs to $\mathbf{Min}_{\psi^3} \mathbf{CoLearn}$.

(D) There exists infinitely many $j \in N$ such that, $\text{card}(\{\psi_r^3 \mid r \leq g^3(j) \wedge \psi_r^3 \in \mathcal{C}^3\}) > 2j$.

Proof. Let

$$f_i(x) = \begin{cases} 3 & \text{if } x = 0, \\ i & \text{if } x = 1, \\ 0 & \text{otherwise.} \end{cases}$$

Let $h(0) = 0$; $h(i + 1) = h(i) + i + 1$.

For $h(i) \leq j < h(i + 1)$, let ψ_j^3 be defined in such a way that the following two properties are satisfied:

(a) $\psi_j^3 \subseteq f_i$;

(b) ψ_j^3 is total iff $(\forall m < i)[F_{\mathbf{M}_m}(A, f_i) \cap \{y \mid y < h(i + 1)\} \neq \{y \mid y < h(i + 1) \wedge y \neq j\}]$.

Note that one can easily construct such ψ^3 . Let $g^3(k) = h(2k + 1)$.

From the properties of ψ^3 discussed above, it is easy to verify that, for all i , there exists a j , $h(i) \leq j < h(i + 1)$, such that $\psi_j^3 = f_i$. Moreover, for all $m < i$, $f_i \notin \mathbf{Min}_{\psi^3} \mathbf{CoLearn}(\mathbf{M}_m)$. Thus properties (A), (C) and (D) of the lemma are satisfied.

Note that in the limit it is easy to verify, whether $(\forall m < i)[F_{\mathbf{M}_m}(A, f_i) \cap \{y \mid y < h(i + 1)\} \neq \{y \mid y < h(i + 1) \wedge y \neq j\}]$. Thus in the limit, for each f_i , one can find the minimum j , such that $h(i) \leq j < h(i + 1)$, and $(\forall m < i)[F_{\mathbf{M}_m}(A, f_i^3) \cap \{y \mid y < h(i + 1)\} \neq \{y \mid y < h(i + 1) \wedge y \neq j\}]$. It follows that $\mathcal{C}^3 \in \mathbf{Min}_{\psi^3} \mathbf{Ex}$. \square

5. Conclusions

In this paper we studied identification by minimal grammars for **FIN**, **CoLearn**, and **Ex**-identification criteria in Kolmogorov and Gödel numberings. We showed that for

every Kolmogorov numbering, ψ , **FINITE**, **Min $_{\psi}$ FIN**, **Min $_{\psi}$ CoLearn**, and **Min $_{\psi}$ Ex** are distinct. We also showed that every possible relationship consistent with **FINITE** \subseteq **Min $_{\psi}$ FIN** \subseteq **Min $_{\psi}$ CoLearn** \subseteq **Min $_{\psi}$ Ex** can be realized for some Gödel numbering ψ . In addition we compared minimal identification in Kolmogorov numberings vis-a-vis Gödel numberings.

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